```
In [ ]: import pandas as pd
        import numpy as np;
      C:\Users\UMMAR\anaconda3\lib\site-packages\numpy\_distributor_init.py:30: UserWarning: loaded more than 1 DLL fr
      om .libs:
      C:\Users\UMMAR\anaconda3\lib\site-packages\numpy\.libs\libopenblas.EL2C6PLE4ZYW3ECEVIV30XXGRN2NRFM2.gfortran-win
       amd64.dll
      C:\Users\UMMAR\anaconda3\lib\site-packages\numpy\.libs\libopenblas.WCDJNK7YVMPZQ2ME2ZZHJJRJ3JIKNDB7.gfortran-win
       amd64.dll
       warnings.warn("loaded more than 1 DLL from .libs:"
In [ ]: from sklearn.model_selection import train_test_split
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense, Dropout
        from sklearn.metrics import roc_curve
        import matplotlib.pyplot as plt
        from sklearn.metrics import confusion matrix
        import seaborn as sns
In [ ]: df = pd.read_csv("diabetes.csv")
In [ ]: df.isna().sum()
Out[]: Pregnancies
                                    0
                                    0
        Glucose
        BloodPressure
                                    0
        SkinThickness
                                    0
        Insulin
                                    0
        BMT
                                    0
        DiabetesPedigreeFunction
                                    0
        Age
                                    0
        Outcome
                                    0
        dtype: int64
In [ ]: df.shape
Out[]: (768, 9)
In [ ]: for col in df.columns:
                missing\_rows = df.loc[df[col]==0].shape[0]
                print(col + ": " + str(missing_rows))
      Pregnancies: 111
      Glucose: 5
      BloodPressure: 35
      SkinThickness: 227
      Insulin: 374
      BMI: 11
      DiabetesPedigreeFunction: 0
      Age: 0
      Outcome: 500
In [ ]: df['Glucose'] = df['Glucose'].replace(0, np.nan)
        df['BloodPressure'] = df['BloodPressure'].replace(0, np.nan)
        df['SkinThickness'] = df['SkinThickness'].replace(0, np.nan)
        df['Insulin'] = df['Insulin'].replace(0, np.nan)
        df['BMI'] = df['BMI'].replace(0, np.nan)
        df['Glucose'] = df['Glucose'].fillna(df['Glucose'].mean())
        df['BloodPressure'] = df['BloodPressure'].fillna(df['BloodPressure'].mean())
        df['SkinThickness'] = df['SkinThickness'].fillna(df['SkinThickness'].mean())
        df['Insulin'] = df['Insulin'].fillna(df['Insulin'].mean())
        df['BMI'] = df['BMI'].fillna(df['BMI'].mean())
In [ ]: for col in df.columns:
                missing\_rows = df.loc[df[col]==0].shape[0]
                print(col + ": " + str(missing_rows))
      Pregnancies: 111
      Glucose: 0
      BloodPressure: 0
      SkinThickness: 0
      Insulin: 0
      BMI: 0
      DiabetesPedigreeFunction: 0
      Age: 0
      Outcome: 500
In [ ]: df.describe()
```

```
3.845052 121.686763
                                            72.405184
                                                           29.153420 155.548223
                                                                                 32.457464
                                                                                                                      33.240885
         mean
                                                                                                           0.471876
           std
                   3.369578
                             30.435949
                                            12.096346
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          25%
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                                            64.000000
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          50%
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          75%
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          max
In [ ]: from sklearn import preprocessing
In [ ]: df_scaled = preprocessing.scale(df)
         df_scaled = pd.DataFrame(df_scaled, columns=df.columns)
         df scaled['Outcome'] = df['Outcome']
         df = df_scaled
In [ ]: X = df.loc[:, df.columns != 'Outcome']
         y = df.loc[:, 'Outcome']
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state= 13)
         # Build neural network in Keras
         model = Sequential()
         model.add(Dense(64, activation='relu', input_dim=8))
         model.add(Dropout(0.2))
         model.add(Dense(64, activation='relu'))
         model.add(Dropout(0.2))
         model.add(Dense(64, activation='relu'))
         model.add(Dense(64, activation = "relu"))
         model.add(Dense(1, activation='sigmoid'))
         model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
         model.fit(X_train, y_train, epochs=200, verbose=False)
         # Results - Accuracy
        scores = model.evaluate(X_train, y_train, verbose=False)
print("Training Accuracy: %.2f%\n" % (scores[1]*100))
         scores = model.evaluate(X_test, y_test, verbose=False)
         print("Testing Accuracy: %.2f%\n" % (scores[1]*100))
       Training Accuracy: 98.86%
       Testing Accuracy: 72.73%
```

Insulin

768.000000 768.000000 768.000000

Age

768.000000 768.000000 76

BMI DiabetesPedigreeFunction

Out[]:

count

Pregnancies

768.000000 768.000000

In []: (model.predict(X_test) > 0.5).astype("int32")

5/5 [======] - 0s 1ms/step

Glucose BloodPressure SkinThickness

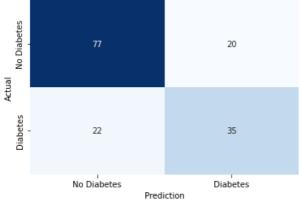
768.000000

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In [ ]: # Results - Confusion Matrix
        y_test_pred = (model.predict(X_test) > 0.5).astype("int32")
      5/5 [======] - 0s 1ms/step
In [ ]: c_matrix = confusion_matrix(y_test, y_test_pred)
        ax = sns.heatmap(c_matrix, annot=True, xticklabels=['No Diabetes', 'Diabetes'], yticklabels=['No Diabetes', 'Diabetes']
```

In []:

```
ax.set_xlabel("Prediction")
ax.set_ylabel("Actual")
plt.show()
plt.clf()
```



<Figure size 432x288 with 0 Axes>

```
In []:

# Results - ROC Curve

y_test_pred_probs = model.predict(X_test)

FPR, TPR, _ = roc_curve(y_test, y_test_pred_probs)

plt.plot(FPR, TPR)

plt.plot([0,1],[0,1],'--', color='black') #diagonal line

plt.title('ROC Curve')

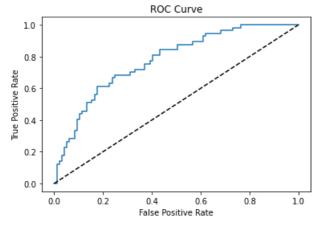
plt.xlabel('False Positive Rate')

plt.ylabel('True Positive Rate')

plt.show()

plt.clf();
```

5/5 [======] - 0s 1ms/step



<Figure size 432x288 with 0 Axes>

